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1 Transferability of mechanistic ecological models is about emergence from first 2 principles

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5 Viktoriia Radchuk¹, Stephanie Kramer-Schadt^{1,2}, Volker Grimm^{3,4,5}
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7 ¹Department of Ecological Dynamics, Leibniz Institute for Zoo and Wildlife Research (IZW), Alfred-
8 Kowalke-Straße 17, Berlin, Germany, radchuk@izw-berlin.de

9 ²Department of Ecology, Technische Universität Berlin, Rothenburgstrasse 12, 12165 Berlin,
10 kramer@izw-berlin.de

11 ³Department of Ecological Modelling, Helmholtz Centre for Environmental Research – UFZ,
12 Permoserstr. 15, Leipzig, Germany, volker.grimm@ufz.de

13 ⁴Institute for Biochemistry and Biology, University of Potsdam, Maulbeerallee 2, Potsdam, Germany

14 ⁵German Centre for Integrative Biodiversity Research (iDiv) Halle-Jena-Leipzig, Deutscher Platz 5e,
15 Leipzig, Germany
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19 Because of the lack of time, data and resources and the need for urgent actions, ecologists often
20 transfer models developed for one study system to a different context. Such transfers imply multiple
21 challenges, which are identified by Yates and colleagues [1]. Although being insightful and elaborate,
22 their review is almost exclusively focusing on correlative species distribution models (SDMs) whereas
23 in their title they refer to “ecological models”, which would also include mechanistic models.
24

25 Some of the issues of transferring correlative and mechanistic models overlap, as pointed out by Yates
26 and colleagues [1] in their Box 3, but some are also unique to mechanistic models and have been
27 identified only over the last 10 years or so. As Yates et al. are writing, traditionally also many
28 mechanistic models were entirely based on empirical, i.e. correlative relationships, but modellers are
29 increasingly replacing imposed, empirical relationships with models in which behaviours emerge from
30 the adaptive decision making of individual organisms, or similar first principles. Thus, one main
31 challenge for the transferability of mechanistic models is estimating the degree to which processes can
32 be imposed vs. should be modelled as emerging property from underlying, first principles.
33

34 Mechanistic ecological models have been transferred on multiple occasions [4, 5] but so far the
35 success is mixed [6, 7]. A main limitation is the legacy of “demographic thinking”, which fails to
36 make the distinction between imposed and emergent mechanisms. Demographic rates, for example
37 mortality, are often used as parameters in population dynamics model and parameterized via, e.g.,
38 mark-recapture studies. In this way mortality is imposed, so that the model reflects the conditions
39 under which the underlying data were collected (Fig. 1). Simply extrapolating the model to new
40 conditions can be highly misleading, as has been shown with model addressing winter mortality of
41 shorebirds [8]. SDMs are facing the same challenge, as pointed out by Yates et al [1].
42

43 To allow transfer to new conditions, any aggregated parameters, like demographic rates or parameters
44 describing species presence-environment relationships, must emerge from what the building blocks of
45 ecological systems, the organisms, are doing (Fig. 1). In other words, the behaviour of the organisms
46 should emerge from first principles such as energy budgets, stoichiometry, photosynthesis, resource

47 uptake, or more generally fitness seeking [9]. A further requirement is to generically capture the
48 interactions among individuals, in particular competition, facilitation, and trophic relationships.
49 Examples of this “next-generation” type of ecological models [9] that allow transfer to new conditions
50 include models of tropical forest growth and dynamics based on photosynthesis and allometric
51 relationships [10], and models of invertebrate population dynamics based on Dynamic Energy Budget
52 theory [11].

53

54 Consequently, these challenges were not identified for correlative SDMs [1], as relations in such
55 models are exclusively imposed. Further, some of the challenges identified by Yates et al. [1] for
56 correlative SDMs are irrelevant for mechanistic models. For example, the issue of what response
57 variables make a model transferable [1] does not apply to mechanistic models, because what is a
58 response variable in a correlative SDM (abundance or presence-absence) usually emerges from lower-
59 level processes in mechanistic models. Also, the issue of incorporating species interactions in model
60 transfers, identified by Yates et al. [1], is rather naturally dealt with in the context of mechanistic
61 models using the individual as the lowest entity.

62

63 We concur with Yates and colleagues [1] that solving the issues of model transferability requires
64 establishing standards for assessing transferability and investigating the determinants of ecological
65 predictability. We submit that an indispensable way to address some of the transferability issues is by
66 using next-generation mechanistic ecological models that are ideally based on first principles. Such
67 models are more generally applicable, i.e. across systems and closely related species, and thus more
68 transferable. Moreover, mechanistic models may alleviate some of the transferability issues of the
69 static models by generating range dynamics as a property emerging from the underlying population-
70 level processes (as in Dynamic Range Models *sensu* [12]). Ecology needs both, correlative and
71 mechanistic models, and none of them is more important than the other, or should be ignored.

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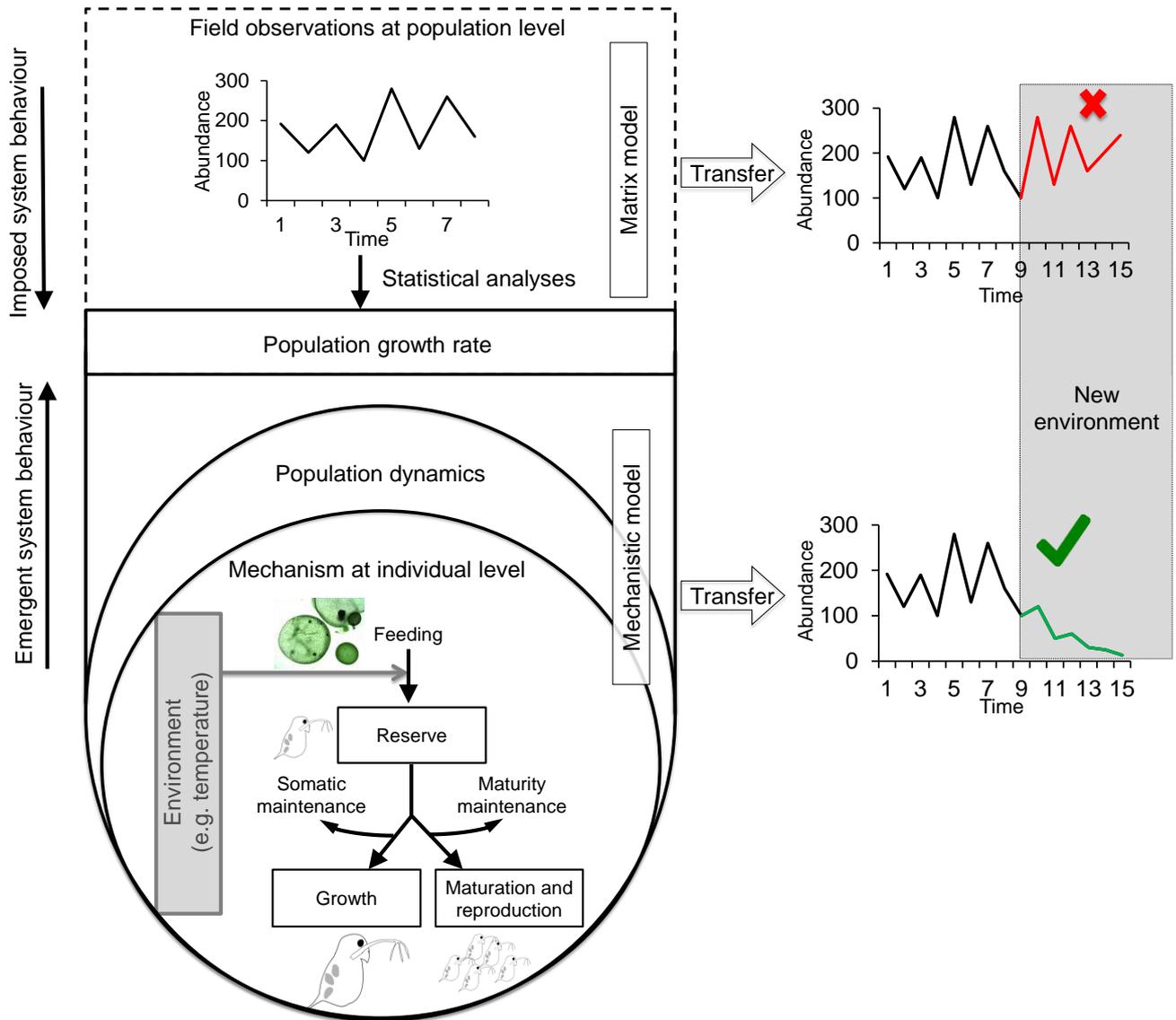


Fig 1. The system behaviour, i.e. here population dynamics, may be imposed by using demographic parameters obtained from statistical analyses of empirical data, e.g. with capture-recapture and survival analyses. This is often done, for example, in population projection matrix models. On the other hand, the system behaviour in dynamic ecological models emerges from lower-level mechanisms at the individual level. The imposed and emergent system behaviours are indicated by a downward and an upward arrow, respectively, shown at the left of the scheme. The models with imposed system behaviour fail to capture the underlying mechanisms and therefore often fail when transferred to new conditions, as shown with the projections of population abundance on the right (incorrectly projected population dynamics in red). On the contrary, the transfers using dynamic mechanistic models are expected to be successful (population dynamics in green).